



Techno-Economic Analysis of Li-ion Batteries in the Capacity Market with Different Degradation Models

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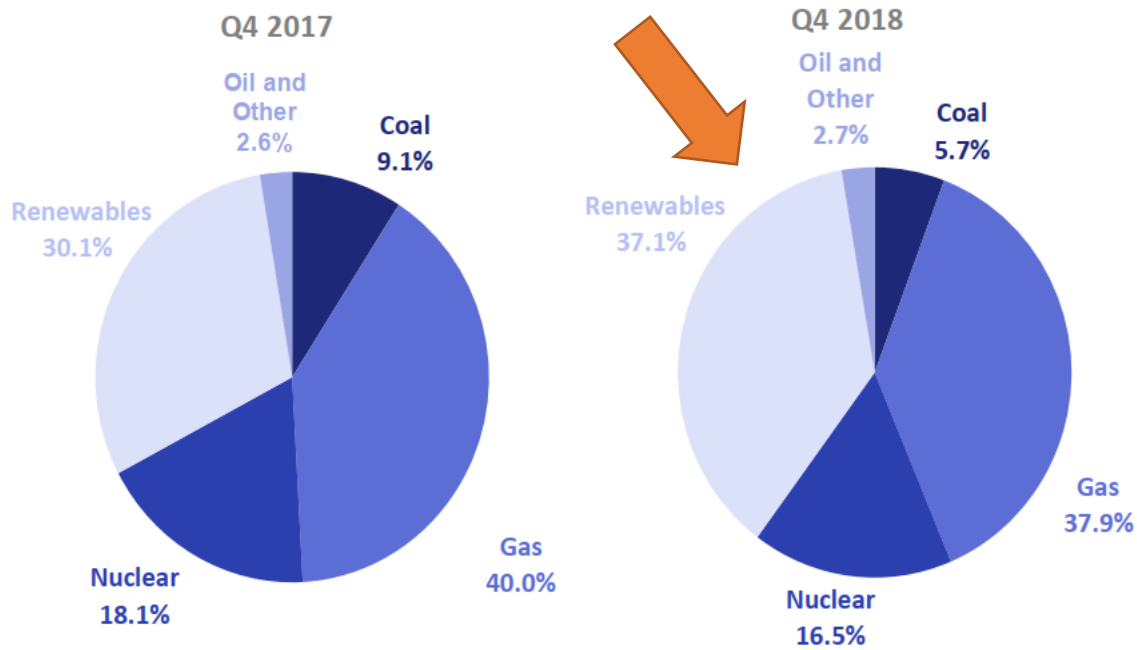
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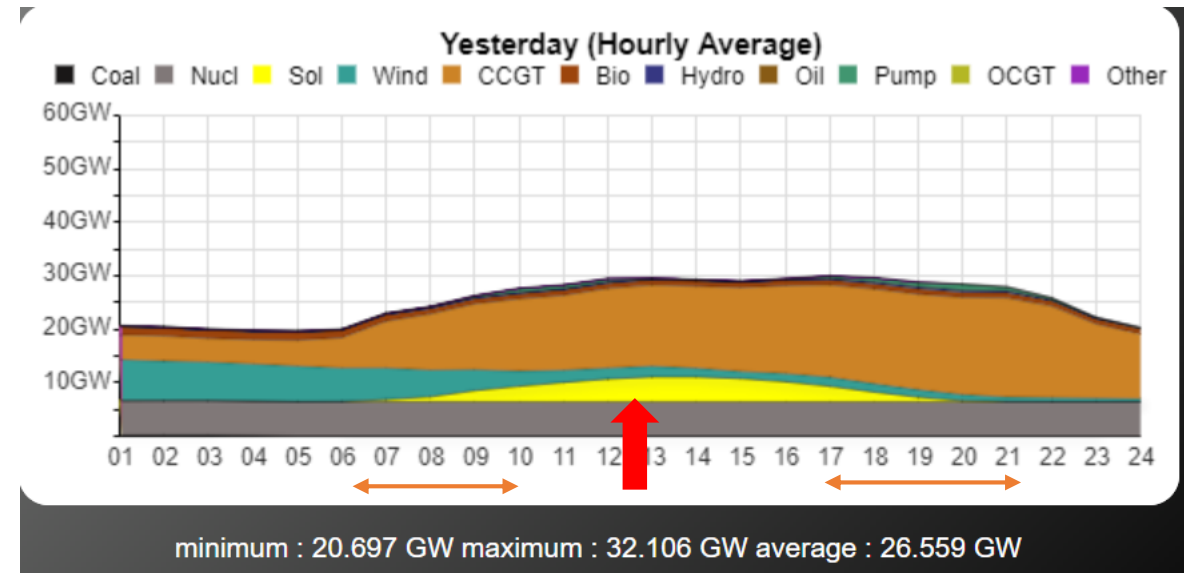
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Introduction



Source: BEIS, Energy trends, 2019



Source: UK Grid watch, 2019

Energy-only markets

- Energy-only market **only compensates generated MW that is actually produced**
- Capacity is only indirectly compensated based on implicit supply agreements, such as futures contracts
- The question is what happens with the prices for imbalances when there is a loss of load occasion (LOLO). If there is a LOLO (i.e., demand larger than available capacity), then this means that there is at least one trader/retailer who sells power without having that power to sell.



Energy-only markets

The Guardian

UK ► UK politics Education Media Society Law Scotland Wales Northern Ireland

National Grid

UK energy watchdog demands answers after major power cut

Outage caused travel chaos and cut electricity to almost 1m people in England and Wales



Energy

How batteries stopped the UK's power cut being a total disaster

On Friday afternoon, Britain's power grid fell apart when two power plants went offline. But a handful of battery firms prevented things from getting much worse

INDEPENDENT

UK power cut: What happened, why it was so bad and who is to blame for outage that left 1 million in the dark?

Incident a 'wake-up call' to energy industry

BBC NEWS

UK power cut: Andrea Leadsom launches government investigation

11 August 2019

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UK power cut

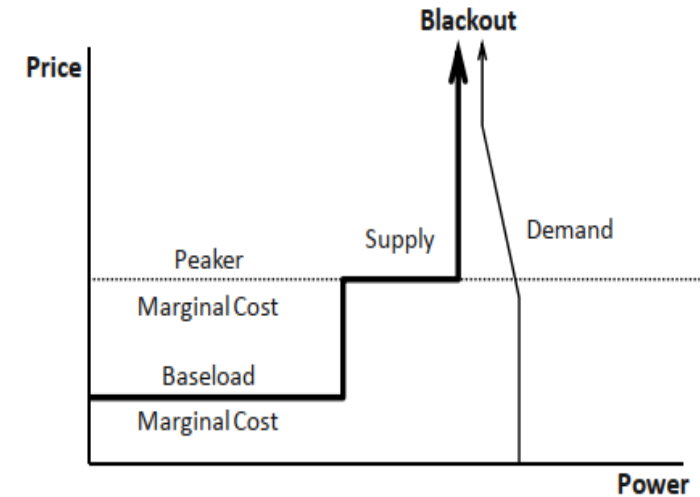
London experienced rush-hour chaos today when the power died across the country

Source: Telegraph

Energy-only market failure



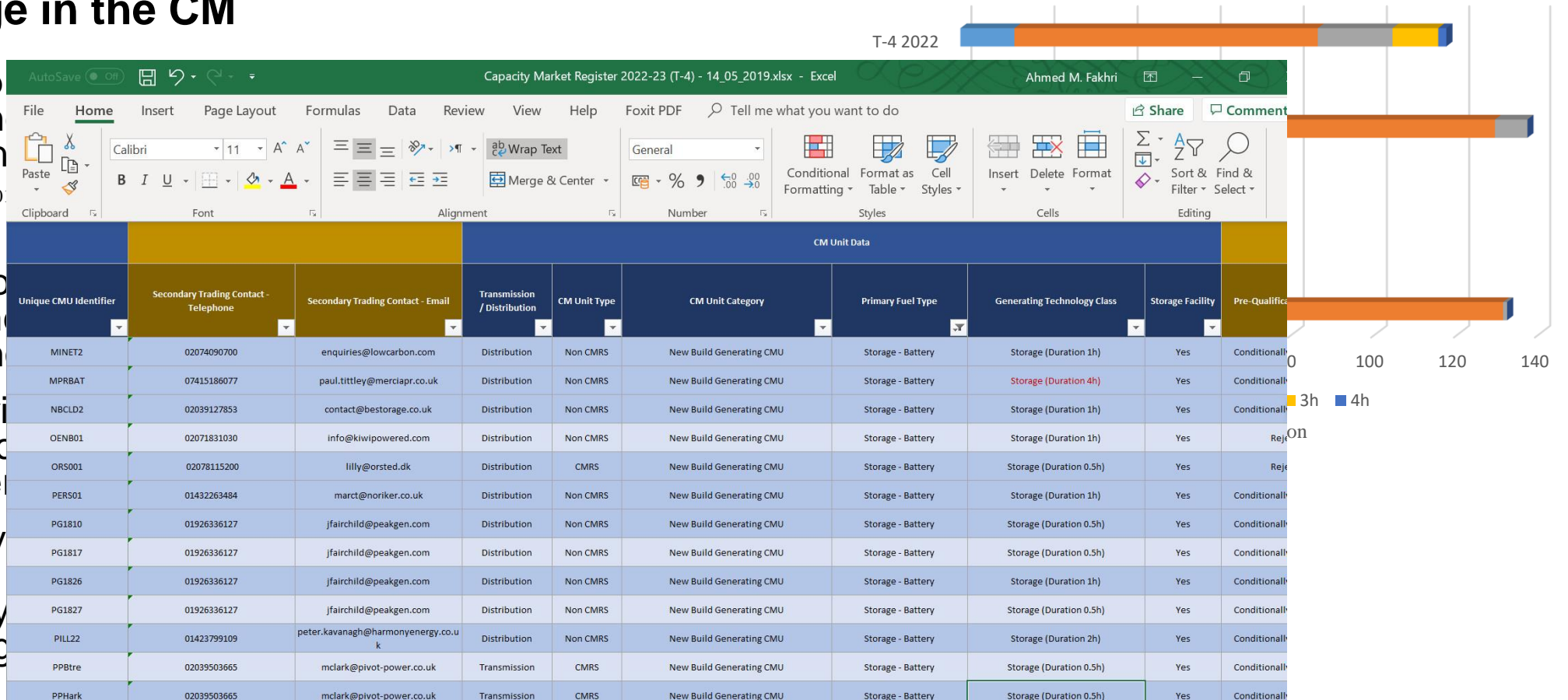
because customers are unaware of the real-time price of electricity or have no reason to respond to them



Source: Capacity market fundamentals, Cramton et.al,2013

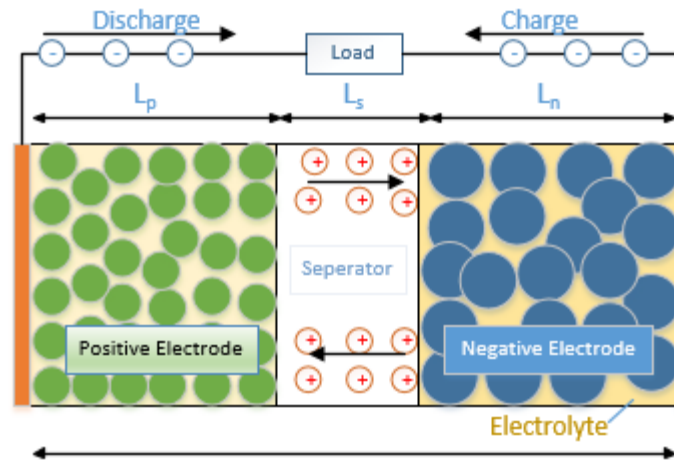
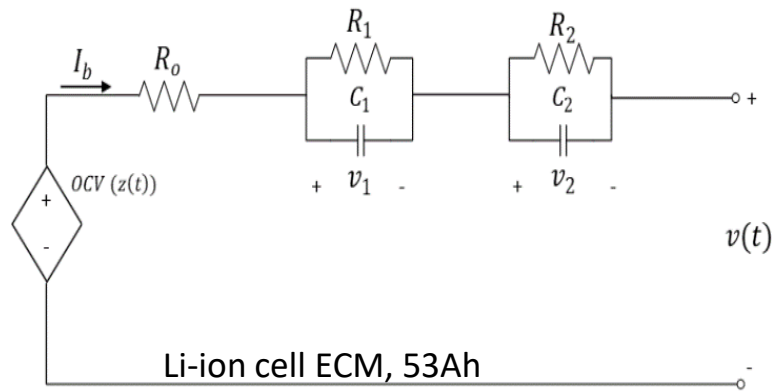
Energy Storage in the CM

- Batteries can provide a viable alternative to traditional generation investment in new plant (S. Forrester, et al. 2018)
- It is found that the case for participation in the CM is reduced by the inclusion of reducing the share of revenue from the CM at all, or commissioning batteries
- Energy delivery time during the CM remain in a fully charged state able to discharge occur (Gissey et.al, 2018)

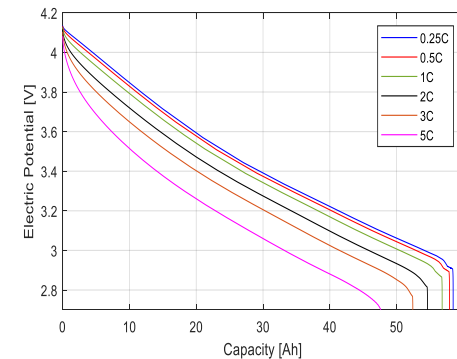
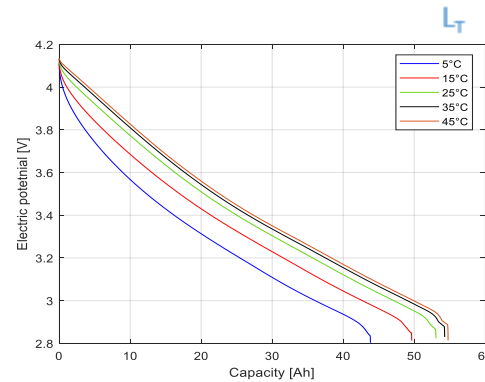
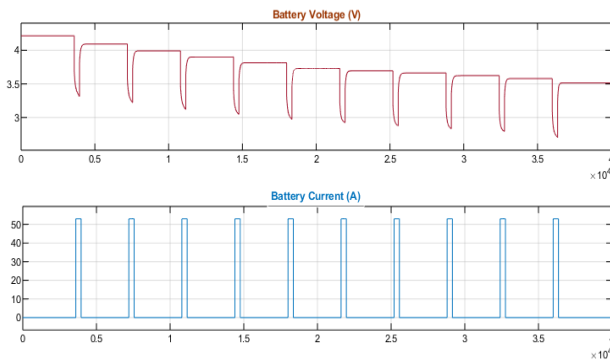


Methods

1- Battery models:



Battery	
Cn = 53Ah, Vn = 3.7V	
String: 12 cells in series, 3.7*12 = 44.4V @53Ah	
Module: 4 strings in parallel, 4*53=212Ah @44.4V	
Rack = 14 modules in series, 15*44.4 = 666V @212Ah	
Battery = 15 rack in parallel, 14*212=2968Ah @666V	
Total: 2968*666 ≈ 2 MWh	
Number of cells (N) ≈ 10080	



Methods

1- Battery models:

Volume fraction of the active material, ϵ_s	0.58 [22]		0.43 [22]
Volume fraction of the electrolyte, ϵ_e	0.32 [22]	0.54 [22]	0.32 [22]
Maximum lithium concentration in the solid phase, $C_{s, \max}$ (mol m^{-3})	29802 ^b		87593 ^b
Electrolyte lithium concentration, $C_{e, \max}$ (mol m^{-3})		1200 [22,27]	
Stoichiometry at SOC = 1, x_1, y_1	0.75 ^b		0.38 ^b
Stoichiometry at SOC = 0, x_0, y_0	0.05 ^b		0.93 ^b
R_{SEI} (Ωcm^2)	20		20
Lithium ion transference number, t_+^0	0.363	0.363	0.363
Electrolyte activity coefficient, f_{\pm}	1 [27]	1 [27]	1 [27]
Charge transfer coefficient, α	0.5 [27]		0.5 [27]
Dynamic parameters			
Lithium diffusion coefficient in the negative electrode, $D_{s, \text{neg}}$ ($\text{m}^2 \text{s}^{-1}$)	$D_{s, \text{neg}} = 3 \times 10^{-13} \exp\left(\frac{-E D_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$	(22)	
Lithium diffusion coefficient in the positive electrode, $D_{s, \text{pos}}$ ($\text{m}^2 \text{s}^{-1}$)	$D_{s, \text{pos}} = 7 \times 10^{-14} \exp\left(\frac{-E D_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$	(23)	
Lithium diffusion coefficient in the electrolyte, D_e ($\text{m}^2 \text{s}^{-1}$)	$D_e = 3.8037e - 10 \times \exp(-0.78281 C) \exp\left(\frac{-E D_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$ [61]	(24)	
Reaction rate in the negative electrode, k_{neg} (m s^{-1})	$k_{\text{neg, discharge}} = k_{0, \text{neg}}^{\text{dis}} \exp\left(\frac{-E k_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$ $k_{\text{neg, charge}} = k_{0, \text{neg}}^{\text{ch}} \exp(-5x) \exp\left(\frac{-E k_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$	(25)	
Reaction rate in the positive electrode, k_{pos} (m s^{-1})	$k_{\text{pos, discharge}} = k_{0, \text{pos}}^{\text{dis}} \exp(-5y) \exp\left(\frac{-E k_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$ $k_{\text{pos, charge}} = k_{0, \text{pos}}^{\text{ch}} \exp\left(\frac{-E k_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$	(26)	
Electrolyte ionic conductivity, κ (S cm^{-1})	$\kappa = 15.8 \times c_e \cdot \exp(-13472 C e^{1.4}) \exp\left(\frac{-E \kappa_0}{R} \left(\frac{1}{T} - \frac{1}{298.15}\right)\right)$ [27]	(27)	
Open circuit potential of the negative electrode	$U_{\text{ref, neg}} = 0.6379 + 0.5416 \exp(-305.5309x) + 0.044 \tanh\left(\frac{-x - 0.1958}{0.1088}\right)$ $- 0.1978 \tanh\left(\frac{x - 0.0117}{0.0529}\right) - 0.0175 \tanh\left(\frac{x - 0.5692}{0.0875}\right)$ [62]	(28)	
Open circuit potential of the positive electrode	$U_{\text{ref, pos}} = -10.72y^4 + 23.88y^3 - 16.77y^2 + 2.595y + 4.563$ [63]	(29)	
Local state of charge of the negative electrode, x	$x = \text{SOC}_{\text{neg}} = \frac{C_{s, \text{surf, neg}}}{C_{s, \text{max, neg}}}$	(30)	
Local state of charge of the positive electrode, y	$y = \text{SOC}_{\text{pos}} = \frac{C_{s, \text{surf, pos}}}{C_{s, \text{max, pos}}}$	(31)	

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A systematic approach for electrochemical-thermal modelling of a large format lithium-ion battery for electric vehicle application

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^aWMG, University of Warwick, Coventry CV4 7AL, United Kingdom



2- Battery degradation models

A-Empirical (E) by Schmalstieg et.al

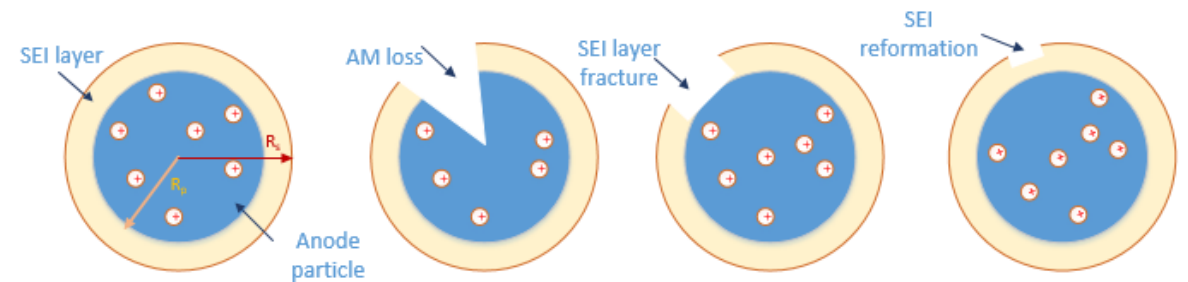
$$C_{lost1} = (\alpha(V_{av}, T)t^{0.75} + \beta(DoD, V_{av}) \times 2DoDN_c C_n \int I(t) dt)$$

B-Semi-Empirical (S) by NREL

$$C_{lost2} = \min(Q_{li}(t, T, N_c, DoD, V_{oc}(t)), Q_{neg}(T, DoD, N_c), Q_{pos}(DoD))$$

C-Physics (P) by Jin and Liu

$$C_{lost3} = \sum \left(\int_0^t i_s(t) dt + \int_0^t SoC d\varepsilon_{AM} + \sum_{i=1}^{N_c} Q_{SEI,crack} \right)$$



3- CM regulations and scenarios

Battery Capacity	Connection Capacity	De-rated capacity	Hours
2 MWh	2 MW	0.43 MW	0.5
2 MWh	2 MW	0.81 MW	1
2 MWh	1 MW	0.68 MW	2
2 MWh	0.5 MW	0.48 MW	4

The Revenue (R) of the battery in the CM is calculated as:

$$R = C_{de} \lambda_{cl} f + R_{ov} - p$$

Where

$$C_{de} = P_c k_{de},$$

$$P_c = I_b V_b N,$$

$$p = \sum_{i=1}^n C_{un(i)} \frac{\lambda_{cl}}{24},$$

$$R_{ov} = C_{ov(i)} \min\left(\lambda, \frac{p_T}{C_{ovT}}\right)$$

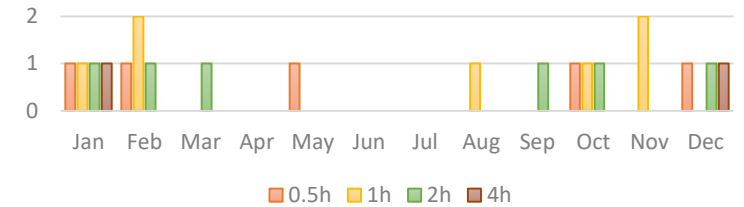
The degradation cost is

$$E_{lost} = C_{lost} \lambda_{degr} N$$

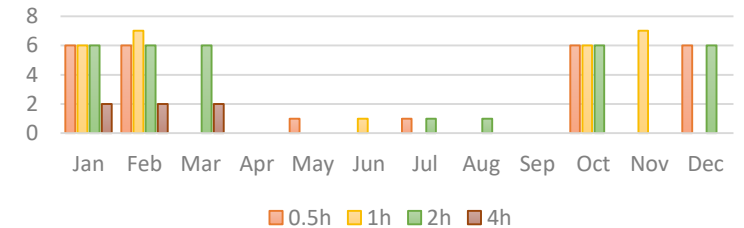
The capacity obligation (C_o) during stress event is calculated:

$$C_o = \sum_{i=1}^n (C_{de} D_p(i)) - C_b(i), \quad D_p = \frac{D_p^{sse}}{C_{ac}}$$

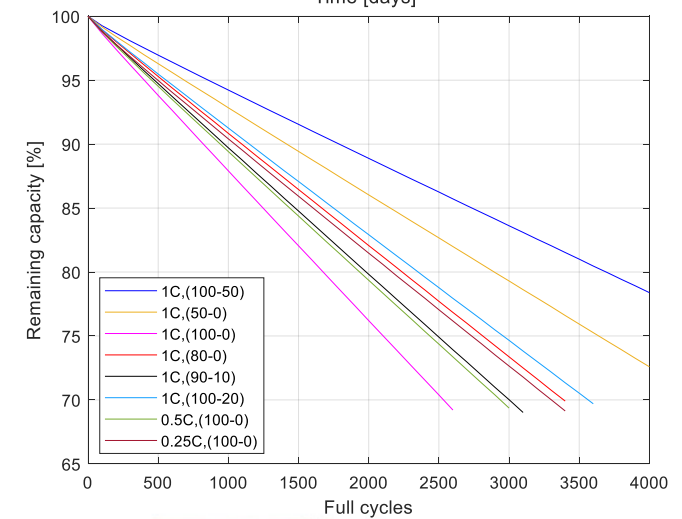
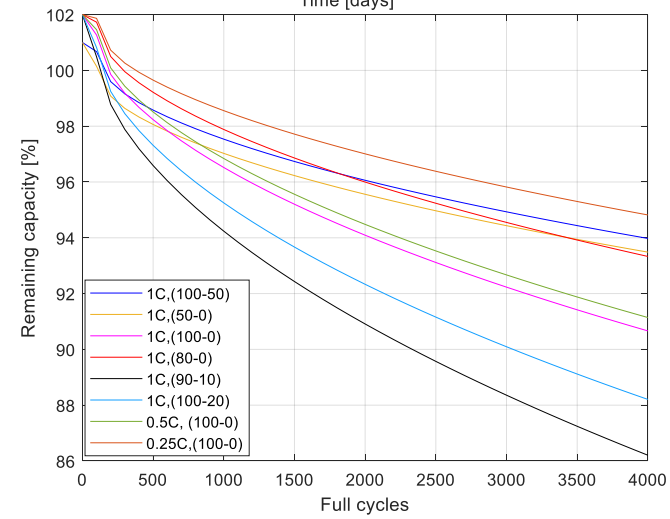
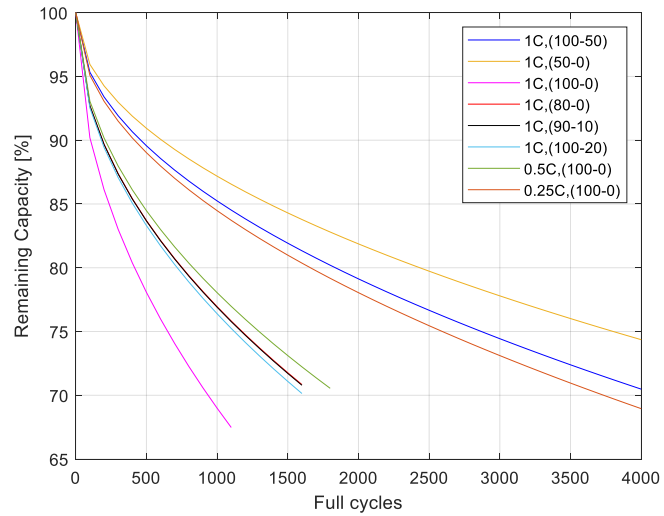
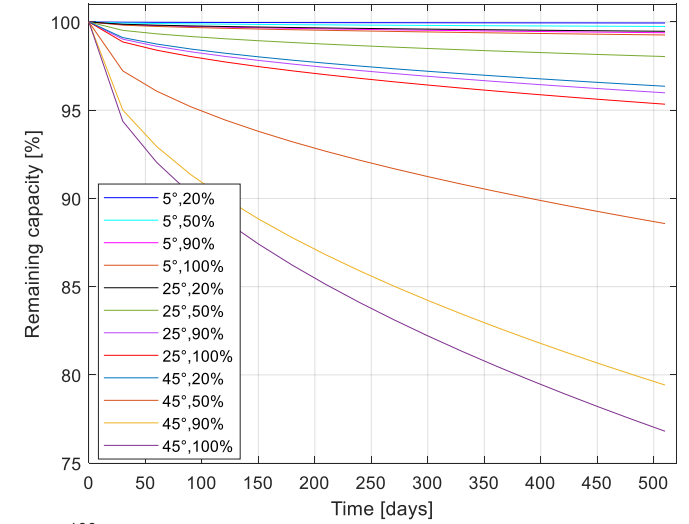
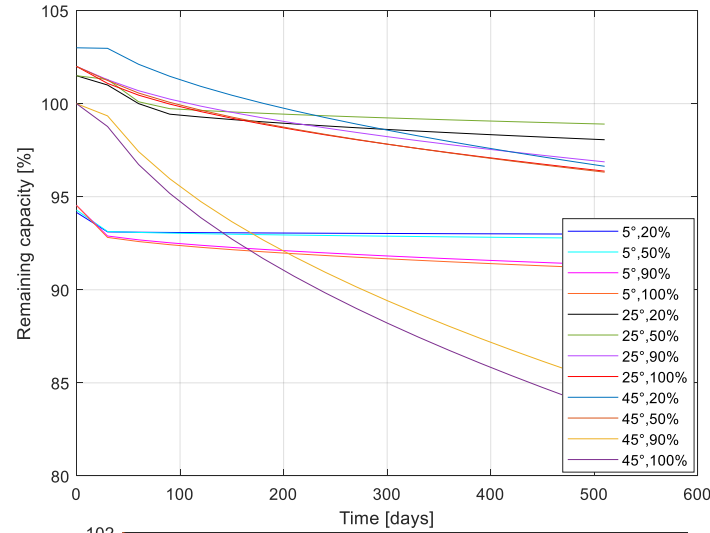
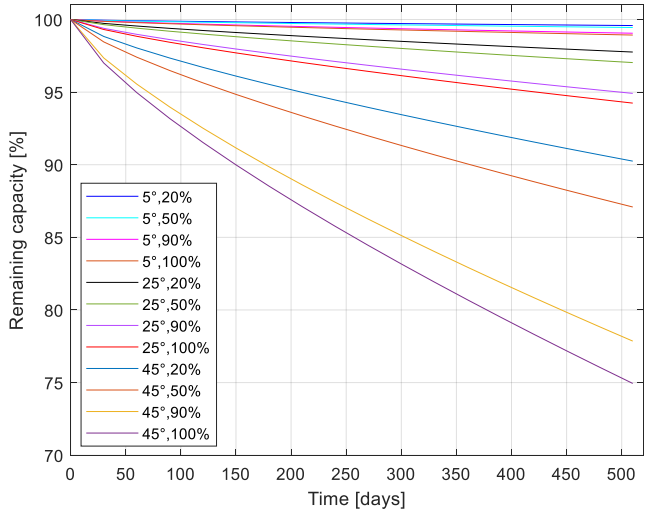
Scenario 1: System Stress Duration Events During One Year



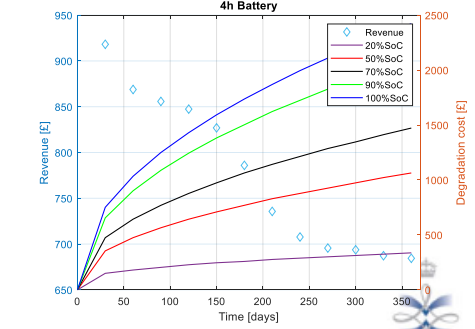
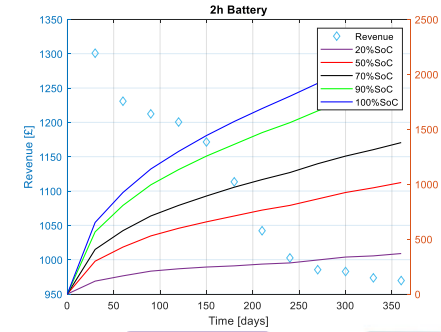
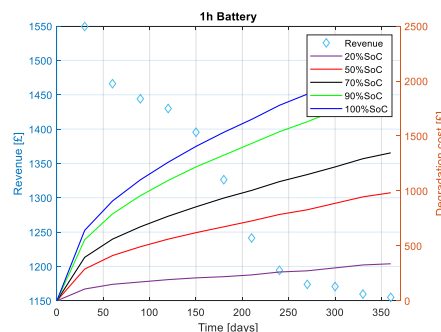
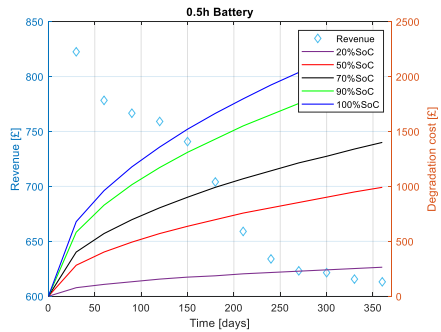
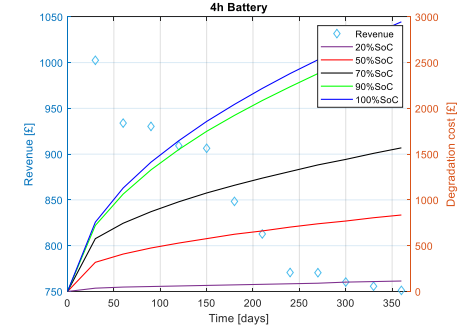
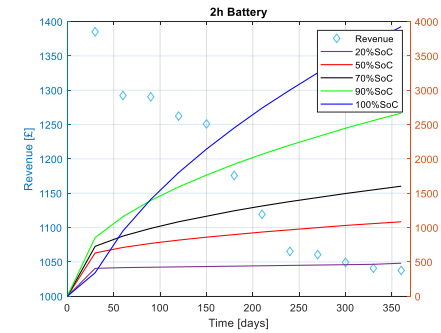
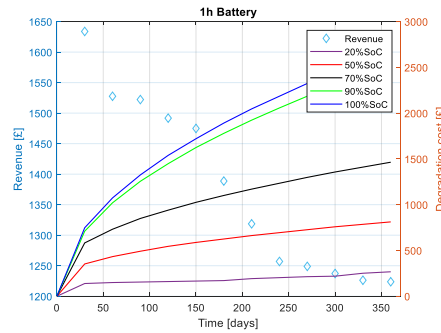
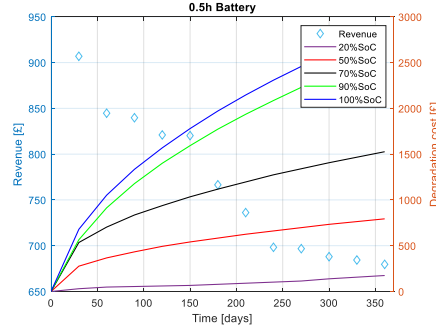
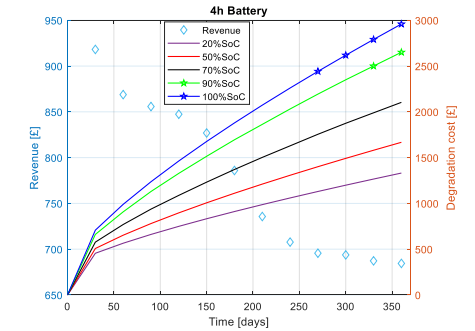
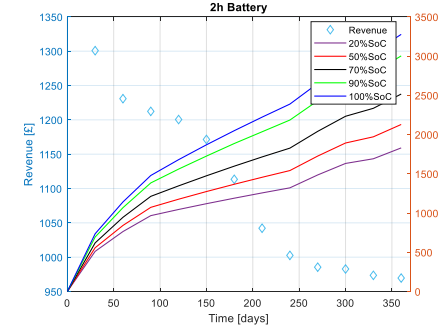
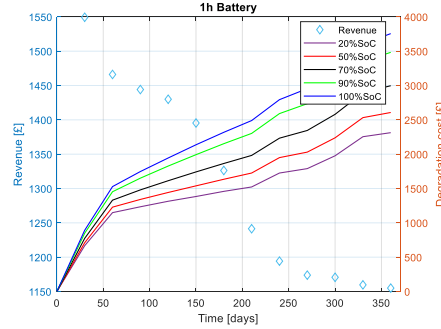
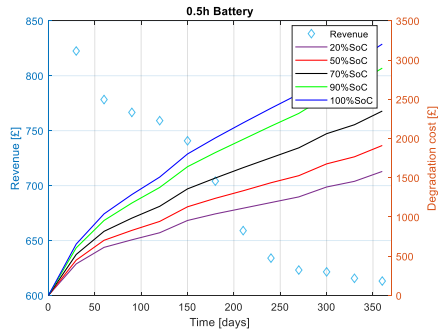
Scenario 2: System Stress Duration Events During One Year



Results



Results



Conclusion

- In general, the results illustrate that degradation cost can significantly impact the potential profit from each battery
- The battery with the 1h de-rating factor shows the highest revenue within the current CM regulations
- The empirical degradation model is simple but overestimate the capacity loss especially in the first cycles. It also underestimates the degradation at low temperatures. Keeping the temperature at 5°C and at low SoC (20%) offers the highest profit
- In contrast, the semi-empirical model shows that the degradation cost is maximum at 5°C and minimal at 25°C with SoC(20%). It shows also a slight increase in the revenue for all cases because the battery capacity predicted shows an increase above nominal in the first 50 days of cycling
- The physics model offer a deeper understanding for the complex degradation mechanisms inside the battery but it also underestimate the degradation at low temperatures.
- Although the three models use the same battery chemistry data (NMC), but this does not necessarily means that extrapolation is possible outside the characterisation setup set when collecting the data



Thank you

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